UNCLASSIFIED

AD 400'398

Reproduced by the

ARMED SERVICES TECHNICAL INFORMATION AGENCY
ARLINGTON HALL STATION
ARLINGTON 12, VIRGINIA



UNCLASSIFIED

MOTICE: When government or other drawings, specifications or other data are used for any purpose
other than in connection with a definitely related
government procurement operation, the U. S.
Government thereby incurs no responsibility, nor any
obligation whatsoever; and the fact that the Government may have formulated, furnished, or in any way
supplied the said drawings, specifications, or other
data is not to be regarded by implication or otherwise as in any manner licensing the holder or any
other person or corporation, or conveying any rights
or permission to manufacture, use or sell any
patented invention that may in any way be related
thereto.

APPLIED MATHEMATICS AND STATISTICS LABORATORIES STANFORD UNIVERSITY CALIFORNIA

GENERALIZED LEAST SQUARES ESTIMATORS FOR RANDOMIZED FRACTIONAL REPLICATION DESIGNS

BY S. ZACKS

TECHNICAL REPORT NO. 86

March 15, 1963

PREPARED UNDER CONTRACT Nonr-225(52)
(NR-342-022)
FOR
OFFICE OF NAVAL RESEARCH



400 398

CENERALIZED LEAST SQUARES ESTIMATORS FOR RANDOMIZED FRACTIONAL REPLICATION DESIGNS

ру

S. Zacks

TECHNICAL REPORT NO. 86

March 15, 1963

PREPARED UNDER CONTRACT Nonr-225(52)

(NR-342-022)

FOR

OFFICE OF NAVAL RESEARCH



Reproduction in Whole or in Part is Permitted for any Purpose of the United States Government

APPLIED MATHEMATICS AND STATISTICS LABORATORIES

STANFORD UNIVERSITY

STANTORD, CALIFORNIA

Generalized Least Squares Estimators for Randomized Fractional Replication Designs

bу

S. Zacks

1. Introduction

Fractional replication designs have become of great importance, especially for industrial experimentation. A missile whose operation is affected simultaneously by dozens of interacting factors would produce a full factorial experiment of impractical size. Indeed, if there are more than 20 factors which may affect the operation of a missile, and if we like to attain complete information on all the main effects and interactions of the controllable factors we would run the experiments over more than $2^{20} = 1,048,576$ treatment combinations. Fractional replication designs are planned to attain information about some of the main effects and interactions by a relatively small number of trials. If the operation of a missle can be controlled with some information on the main effects and some low order linear interations, it might be sufficient to run only 32 or 64 trials at a time. These however, should be chosen from those possible in some optimal manner.

The problem of choosing a 1/2^{m-s} fractional replication and an appropriate estimator of the parameters characterizing the factorial model (main effects and interactions) has been studied by A. P. Dempster (1960, 1961), K. Takeuchi (1961), S. Ehrenfeld and S. Zacks (1961, 1962), S. Zacks (1962), B. V. Shah and O. Kempthorne (1962a,b). In all these studies the type of estimators considered is that which yields, under a randomized procedure with equal probabilities of choice, unbiased

estimates of a specified linear functional of a subvector of parameters, which lies in the range of the design matrix (the matrix of the corresponding normal equations).

In the present study statistical properties of the generalized least squares estimators, under randomized fractional replication designs, are studied. The term generalized least-squares estimators (denoted henceforth by g.l.s.e.) is used since the matrices of the normal equations corresponding to these designs are singular. The factorial models corresponding to the type of fractional replication designs studied in the present paper is presented in section 2. For this sake we start from the factorial model for a full factorial system. Then we present the required algebra, and the method of constructing the orthogonal fractional replications. The linear spaces of all g.l.s.e. associated with the various orthogonal fractional replication designs are characterized in terms of the linear coefficients of the corresponding factorial models. Some statistical properties of the g.l.s.e. under procedures of choosing a fractional replication at random, are then studied. First we prove that there is no g.l.s.e. which yields unbiased estimates of the entire vector of 2^m parameters. However, there are g.l.s.e.s which estimate unbiasedly subvectors of parameters. The trace of the mean-square-error matrix corresponding to a g.l.s.e. applied under certain randomization procedure is used as a loss function for the decision problem of choosing a g.l.s.e. and a randomization procedure. It is shown that when the parameters of the factorial system may assume arbitrary values, the randomization procedure which assigns equal probabilities to various fractional

replications (denoted by R. P.*) is <u>admissible</u>. Bayes g.l.s.e., relative to a-priori information available on the parameters, are then studied. This leads to a minimax theorem, which specifies a minimax and admissible g.l.s.e. under R.P.*.

The relationship between the generalized inverse of the matrix of normal equations and g.l.s.e. as given by A. Ben-Israel and J. Wersen (1962), and by C. R. Rao (1962) is studied. It is shown that these are particular cases in the general class of g.l.s.e. studied presently.

Finally, it should be remarked that although the present paper deals with factorial system of order 2^m all the important results hold in more general factorial systems of order p^m (p>2).

2. The statistical model for fractional replication designs.

2.a. The statistical model for a full factorial experiment of order 2^m.

A full factorial experiment of order 2^m is a set of 2^m treatment combinations, consisting of m factors X_0, \ldots, X_{m-1} each at two levels. Such a system can be characterized by 2^m parameters $\alpha_0, \ldots, \alpha_{m-1}$, which are the coefficients of the multilinear regression function:

$$(2.1) \quad E(Y(X_0, ..., X_{m-1})) = \sum_{(\lambda_0, ..., \lambda_{m-1})} \alpha_{u(\lambda_0, ..., \lambda_{m-1})} x_0^{\lambda_0} x_1^{\lambda_1} ... x_{m-1}^{\lambda_{m-1}}$$

where
$$\lambda_{j} = 0, 1 \ (j=0,...,m-1); \ u(\lambda_{0},...,\lambda_{m-1}) = \sum_{j=0}^{m-1} \lambda_{j} \ 2^{j};$$

and $Y(X_0, ..., X_{m-1})$ is a random variable representing the "yield" of the experiment at treatment combination $(X_0, ..., X_{m-1})$. Denote by $X_{j,0}$ and $X_{j,1}$ (j=0,...,m-1), $X_{j,0} < X_{j,1}$, the two specified levels of factor X_j . By changing variables according to the transformation

(2.2)
$$Z_{j,k} = \frac{X_{j,k} - \frac{1}{2} (X_{j,0} + X_{j,1})}{\frac{1}{2} (X_{j,1} - X_{j,0})}$$
 (k=0,1;j=0,...,m-1)

the regression function (2.1) is reduced to the form:

(2.3)
$$E\{Y(Z_0,...,Z_{m-1})\} = \sum_{u=0}^{2^{m}-1} \beta_u Z_0^{\lambda_0} Z_1^{\lambda_1} ... Z_{m-1}^{\lambda_{m-1}}$$

where $Z_j = -1$, +1 (j=0,...,m-1); and $u=u(\lambda_0,...,\lambda_{m-1})$.

Writing $Z_j = (-1)^{1-ij}$ with $i_j = 0,1$ for all j = 0,...,m-1, the regression function (2.3) can be represented in the form:

(2.4)
$$E\{Y(i_0,...,i_{m-1})\} = \sum_{u=0}^{2^{m}-1} \beta_u(-1)^{\sum_{j=0}^{m-1} \lambda_j(1-i_j)}$$

Furthermore, denote by $x_v = (i_0, ..., i_{m-1})$, $v = \sum_{j=0}^{m-1} i_j 2^j$, the 2^m treatment combinations of the factorial system under consideration; and let

(2.5)
$$c_{vu}^{(2^m)} = (-1)^{j=0}$$
 $\lambda_j(1-i_j)$, for all $v, u = 0, ..., 2^m-1$

then (2.4) is reduced to the form:

(2.5)
$$E\{Y(x_v)\} = \sum_{u=0}^{2^m-1} \beta_u c_{vu}^{(2^m)}$$
, for all $v = 0, ..., 2^m-1$

Let $Y' = (Y(X_0), ..., Y(X_0))$ be the vector of observations at all the 2^m treatment combinations; and let $\beta' = (\beta_0, ..., \beta_{2^m-1})$ be the vector of parameters of (2.5). Thus, if

$$(c^{(2^m)}) = \|c_{vu}^{(2^m)}\|$$
, $(v,u = 0,...,2^m-1)$,

denotes the matrix of the coefficients of the β 's in (2.5), then the statistical model for a full factorial system can be written as:

$$(2.6) Y = (C^{(2^m)}) \beta + \epsilon$$

where ϵ is a random vector, with $E \epsilon = 0$ and $E \epsilon \epsilon' = \sigma^2 I^{(2^m)}$ $I^{(n)}$ denoting the identity matrix of order n).

2.b. The algebra of factorial experiments.

2.b.l. <u>Properties of the matrices</u> (C^(2m))

The properties of the matrices $(C^{(2^m)})$ will be presented without proofs. For details see S. Ehrenfeld and S. Zacks (1961).

The matrices $(C^{(2^m)})$, m = 1, 2, ..., of (2.6) can be obtained recursively by the following relationship:

(2.7)
$$(c^{(2^m)}) = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \otimes (c^{(2^{m-1})}), m = 1,2,...$$

where $(C^{(O)}) \equiv 1$ (scalar), and where $A \otimes B$ is the Kronecker's direct multiplication of the matrix A by the matrix B from the left, defined as follows: If A is an n x m matrix $\|\mathbf{a_{ij}}\|$, and B is a k x l matrix $\|\mathbf{b_{rs}}\|$, then $A \otimes B$ is the nk x ml matrix:

$$\mathbf{A} \otimes \mathbf{B} \begin{bmatrix} \mathbf{a}_{11} & \mathbf{B} & \cdots & \mathbf{a}_{1m} & \mathbf{B} \\ \vdots & \ddots & \vdots & \vdots \\ \mathbf{a}_{n1} & \mathbf{B} & \cdots & \mathbf{a}_{nm} & \mathbf{B} \end{bmatrix}$$

From the associative property of the Kronecker's direct multiplication it follows that every matrix $(C^{\binom{2^m}{}})$ can be factorized into $2^{m-s} \times 2^{m-s}$ $(1 \le s \le m)$ submatrices of order $2^s \times 2^s$, according to the relationship:

(2.8)
$$(c^{(2^m)}) = (c^{(2^{m-s})}) \otimes (c^{(2^s)}), (1 \le s \le m)$$

From this relationship it follows that the elements of $(C^{(2^m)})$ are related to those of $(C^{(2^{m-s})})$ and $(C^{(2^s)})$ according to:

(2.9)
$$C_{1+j2^{s},t}^{(2^{m})} = C_{1,r_{t}}^{(2^{s})} \cdot C_{j,q_{t}}^{(2^{m-s})}$$

for all $i = 0, ..., 2^{8}-1$; $j = 0, ..., 2^{m-8}-1$; and where $t = r_{t} + q_{t}2^{8}$ $(r_{t} = 0, ..., 2^{8}-1; q_{t} = 0, ..., 2^{m-8}-1)$.

Another useful relationship among the elements of $(C^{(2^m)})$ is the following one: Let $u_1, u_2 = 0, \dots, 2^m-1$ be given by $u_1 \equiv (\lambda_{01}, \lambda_{11}, \dots, \lambda_{(m-1)1})$ (i=1,2); $\lambda_{j,1} = 0,1$ $(j=0,\dots,m-1)$, and define $u_1 \oplus u_2 \equiv (\lambda_0',\dots,\lambda_{m-1}')$ where $\lambda_j' \equiv \lambda_{j1} + \lambda_{j2}$ (mod. 2); then

(2.10)
$$c_{\mathbf{v},\mathbf{u}_{1}\oplus\mathbf{u}_{2}}^{(2^{m})} = c_{\mathbf{v},\mathbf{u}_{1}}^{(2^{m})} \cdot c_{\mathbf{v},\mathbf{u}_{2}}^{(2^{m})}$$
 for every $\mathbf{v}=0,\ldots,2^{m}-1$.

The properties of $(C^{(2)})$ are extended into $(C^{(2^m)})$ by the recursion relationship (2.7), and are summarized as follows:

(i)
$$C_{v,0}^{(2^m)} = 1$$
 for every $v = 0,...,2^m-1$

(ii)
$$C_{2^{m}-1,u}^{(2^{m})} = 1$$
 for every $u = 0,...,2^{m}-1$

(iii)
$$\sum_{n=0}^{2^{m}-1} C_{vn}^{(2^{m})} = 0 \text{ for every } v = 0,...,2^{m}-2$$

(iv)
$$\sum_{v=0}^{2^{m}-1} C_{vu}^{(2^{m})} = 0$$
 for every $u = 1, ..., 2^{m}-1$ for every $u = 1, ..., 2^{m}-1$.

$$(v) \sum_{v=0}^{2^{m}_{1}} c_{vu_{1}}^{(2^{m})} c_{vu_{2}}^{(2^{m})} = \begin{cases} 2^{m}, & \text{if } u_{1} = u_{2} \\ \\ 0, & \text{if } u_{1} \neq u_{2} \end{cases}$$

and

$$(vi) \sum_{u=0}^{2^{m}-1} c_{v_{1}u}^{(2^{m})} c_{v_{2}u}^{(2^{m})} = \begin{cases} 2^{m}, & \text{if } v_{1} = v_{2} \\ \\ 0, & \text{if } v_{1} \neq v_{2} \end{cases}$$

(v) and (vi) can be expressed also in the form:

$$(c_{(S_m)}), (c_{(S_m)}) = (c_{(S_m)})(c_{(S_m)}), = S_m I_{(S_m)}$$

2.b.2. The group of parameters.

Every parameter β_u (u = 0,...,2^m-1) of the statistical model (2.6) can be represented by an m-tuple $\beta_u \equiv (\lambda_0, \ldots, \lambda_{m-1})$ where $\lambda_j = 0,1$ (j = 0,...,m-1).

The set of all 2^m parameters constitutes a group, B with respect to the operator \bigotimes , defined as follows:

Let
$$u_1 = \sum_{j=0}^{m-1} \lambda_j 2^j$$
 and $u_2 = \sum_{j=0}^{m-1} \lambda_j^i 2^j$ then $\beta_k \equiv \beta_{u_1} \bigotimes \beta_{u_2}$

if, and only if $k = \sum_{j=0}^{m-1} \lambda_j'' 2^j$ where $\lambda_j'' \equiv \lambda_j + \lambda_j' \pmod{2}$ for all j = 0, ..., m-1.

The unit element of the group B is $\beta_o \equiv (0,0,\ldots,0)$ and the inverse of $\beta_u \equiv (\lambda_0,\ldots,\lambda_{m-1})$ is $\beta_u^{-1} \equiv (2-\lambda_0,2-\lambda_1,\ldots,2-\lambda_{m-1}) \pmod{2}$. A set of n parameters β_{u_1} , β_{u_2} ,..., β_{u_n} is called <u>dependent</u> if there

exist n constants a_k (k=1,...,n), not all of which are zero $(a_k = 0,1)$, such that:

where
$$\left[\beta_u\right]^a = \begin{cases} \beta_u & \text{, if } a=1\\ \beta_0 & \text{, if } a=0 \end{cases}$$

Ť

If relationship (2.11) is valid only when all $a_k = 0$ (k=1,...,n) then $\beta_{u_1},...,\beta_{u_n}$ are called <u>independent</u>. It is easy to check that every set of n independent parameters $(1 \le n \le m)$ generates a subgroup of order 2^n in B.

2.c. The construction of a $1/2^{m-s}$ (1 \leq s , m) fractional replication.

Let $\{\beta_d, \beta_{d_1}, \dots, \beta_{d_{m-s-1}}\}$ be any set of m-s independent parameters in B. The 2^m treatment combinations can be classified into 2^{m-s} disjoint subsets $X_v(v=0,\dots,2^{m-s}-1)$ of equal size, relative to the specified m-s independent parameters, in the following manner: Let $\beta_d \equiv (\lambda_0, d_k, \lambda_1, d_k, \dots, \lambda_{m-1}, d_k)$ $k=0,\dots,m-s-1$ be a defining parameter, and let $X \equiv (i_0,\dots,i_{m-1})$ be a treatment combination, then $x \in X_v$ if, and only if,

(2.12)
$$\sum_{j=0}^{m-1} \mathbf{1}_{j} \lambda_{j,d_{k}} \equiv \mathbf{a}_{k} \pmod{2}, \text{ and where}$$

$$\mathbf{v} = \sum_{k=0}^{m-s-1} \mathbf{a}_{k} 2^{k}.$$

In order to perform the classification of treatment combinations into the blocks X_v ($v=0,\ldots,2^{m-8}$ -1) we do not have to solve the linear equations (2.12), but it suffices to compare the rows of the matrix of coefficients ($C^{(2^m)}$) under the columns corresponding to the special independent parameters (β_d ,..., β_d). These two procedures are equivalent (see S. Ehrenfeld and S. Zacks (1961)). The m-s independent parameters, relative to which the classification takes place, are called defining parameters. The answer to the question, which of the parameters should be specified for the role of defining ones depends on the objectives of the experiment. The choice of a set of defining parameters will generally effect the bias of estimators and their variances, and might have other effects on the properties of statistics and procedures (see O. Kempthorne (1952); S. Ehrenfeld and S. Zacks (1961).

The term <u>fractional replication</u>, in its broadest sense, relates to any subset of treatment combination from a full factorial system. K. Takeuchi (1961) considers designs of randomly combined fractional replications. We shall consider in the present paper only fractional replications which consist of one block of treatment combinations, X_v , chosen from the set of 2^{m-s} blocks constructed according to the procedure outlined above. These fractional replications are called <u>orthogonal</u>. A <u>randomized fractional replication procedure</u> is one in which a block X_v is chosen with a probability vector $\xi' = (\xi_0, \dots, \xi_{2^{m-s}1})$

2.d. The statistical model for a 1/2 m-s fractional replication.

Let $\{\beta_d, \ldots, \beta_d \}$ be a set of defining parameters; $\{X_v ; v = 0, \ldots, 2^{m-s}-1\}$ the corresponding blocks of treatment combinations,

and $Y(X_v)$ the vector of observations associated with the 2^s treatment combinations in X_v . The order of the components of $Y(X_v)$ is determined by the standard order of the corresponding x's in X_v , e.g. if $X_v = \{x_0, x_3, x_5, x_6\}$ then $Y(X_v)' = \{Y(X_0), Y(X_3), Y(X_5), Y(X_6)\}$.

Let $\beta^{(0)}' = (\beta_0, \beta_{t_1}, \dots, \beta_{t_2s_{-1}})$ be any specified vector of 2^s parameters independent of the defining parameters (except for the "mean" β_0); with $t_k < t_{k+1}$ for all $k = 1, \dots, 2^s-1$. Let $\{\beta_u^* : u=0, \dots, 2^{m-s}1\}$ be the subgroup of 2^{m-s} parameters, generated by the m-s defining parameters. Define by $\beta^{(u)}$ $(u = 1, \dots, 2^{m-s}-1)$ the vector of 2^s parameters obtained by multiplying each of the components of $\beta^{(0)}$ by β_u^* , i.e., $\beta^{(u)} = (\beta_0 \bigotimes \beta_u^*, \beta_t \bigotimes \beta_u^*, \dots, \beta_{t_2s_{-1}} \bigotimes \beta_u^*)^s$. Then, the statistical model for $Y(X_v)$ can be written in the form:

(2.13)
$$Y(X_{v}) = \sum_{u=0}^{2^{m-s}-1} (P_{vu}^{(2^{s})}) \beta^{(u)} + \epsilon = (P_{v})\beta^{*} + \epsilon$$

where; as proven by S. Ehrenfeld and S. Zacks (1961)

(2.14)
$$(P_v) = (1, b_{v1}^{(2^{m-s})}, ..., b_{v(2^{m-s}-1)}^{(2^{m-s})}) \otimes (P_{v0}^{(2^s)})$$

is a $2^5x(2^m-2^5)$ matrix; $(P_{vO}^{(2^5)})$ is a 2^5x2^5 matrix obtained from $(C^{(2^m)})$, by picking the elements of $(C^{(2^m)})$ corresponding to treatment combinations in X_v and the parameters in $\beta^{(0)}$, and arranging them in the standard order. The scalars $b_{vu}^{(2^{m-5})}$, by which we multiply $(P_{vO}^{(2^5)})$ to obtain $(P_{vU}^{(2^5)})$, are given by the formula:

(2.15)
$$b_{vu}^{(2^{m-s})} = (-1)^{j=0} (1^{j}_{j} (1^{j-L(d_{j})})$$

where
$$v = \sum_{j=0}^{m-s-1} i_j 2^j$$
; $u = \sum_{j=0}^{m-s-1} i_j 2^j$ and $L(d_j) \equiv \sum_{k=0}^{m-1} \lambda_{k,d_j} \pmod{2}$,

for every defining parameter $\beta_{d,j}$; and where $\beta^* = (\beta^{(0)} : \beta^{(1)} : : : \beta^{(2^{m-s}1)})$, and ϵ is a random vector of order 2^s , with $E\epsilon = 0$ and $E\epsilon\epsilon' = \sigma^2 I^{(2^s)}$.

It can be readily proved that, the rows of $(P_{vu}^{(2^s)})$ $(v,u=0,...,2^{m-s}]$, as well as its columns, are orthogonal, i.e.,

$$(2.16) (P_{v_1}^{(2^8)})^{i} (P_{v_1}^{(2^8)}) = (P_{v_1}^{(2^8)}) (P_{v_1}^{(2^8)})^{i} = 2^8 I^{(2^8)}$$

for all $v,u=0,...,2^{m-s}l$; and that a similar property holds for the matrix $(B_{\mathbf{g}}^{(2^{m-s})})$, whose elements are the coefficients $b_{\mathbf{v}u}^{(2^{m-s})}$, defined by (2.15), i.e.,

$$(2.17) (B(2^{m-s})) (B(d_0, ..., d_{m-s-1})) (B(d_0, ..., d_{m-s-1})) = 2^{m-s} I^{(2^{m-s})}$$

for every choice of m-s defining parameters. For the sake of simplifying notation, let $S=2^{8}$ and $M=2^{m-s}$, $N=S\cdot M=2^{m}$. Furthermore, assume, without loss of generality, that the defining parameters are the "main effects" $\{\beta_{S},\beta_{2S},\ldots,\beta_{N/2}\}$ and that $\beta^{(O)}=(\beta_{O},\beta_{1},\ldots,\beta_{S-1})$ ' then, the blocks of treatment combinations are:

(2.18)
$$\{X_{i+vS}; i = 0,...,S-1\}$$
 for all $v = 0,...,M-1$

and the statistical model for $Y(X_{\mathbf{v}})$ is given by:

(2.19)
$$Y(X_{\mathbf{v}}) = [(1, C_{\mathbf{v}1}^{(M)}, ..., C_{\mathbf{v}(M-1)}^{(M)}) \otimes (C^{(S)})] \beta + \epsilon$$

$$= (C^{(S)})\beta^{(O)} + \sum_{u=1}^{M-1} C_{\mathbf{v}u}^{(M)} \beta^{(u)} + \epsilon$$

where
$$\beta^{(u)} = (\beta_{us}, \beta_{1+us}, ..., \beta_{(u+1)s-1})'$$
.

Generalized least squares estimators for fractional replications. 3.a. The set of all least squares estimators.

Given a block of treatment combinations X_v (v=0,...,M-1) and the associated vector of observations $Y(X_v)$, the "normal equations" corresponding to the linear model (2.19) are given by:

(3.1)
$$(C_{x})^{i} (C_{x})\beta = (C_{x})^{i} Y(X_{x}), v=0,...,M-1,$$

where (C_{v}) is the S x N matrix of the coefficients of (2.19), i.e.,

(5.2)
$$(c_{\mathbf{v}}) = (1, c_{\mathbf{v}1}^{(M)}, \dots, c_{\mathbf{v}(M-1)}^{(M)}) \otimes (c^{(S)}).$$

A generalized least squares estimator (g.l.s.e.) of β is any linear operator (L_v), on E^(S) (Eucliden S-space), so that (L_v) is an N x S matrix satisfying the equation:

(3.3)
$$(C_v)'(C_v)(L_v) = (C_v)', v=0,...,M-1$$
.

Let $(L_v)' = ((L_{vo})' : (L_{vl})' : ... : (L_{v(M-1)})')$ where (L_{vu}) (u=0,...,M-1) are square matrices of order S x S. Substituting from (3.2) for (C_v) in 3.3 and decomposing (L_v) as indicated we arrive at the matrix equation:

$$(3.4) \qquad S[(Q^{(M)}) \otimes (I^{(S)})] \qquad \begin{bmatrix} (L_{vo}) \\ \dots \\ (L_{v(M-1)}) \end{bmatrix} = \begin{bmatrix} 1 \\ C_{v1}^{(M)} \\ \vdots \\ C_{v(M-1)} \end{bmatrix} \otimes (C^{(S)})$$

where
$$(Q^{(M)}) = (1,C_{v1}^{(M)},...,C_{v(M-1)}^{(M)})' (1,C_{v1}^{(M)},...,C_{v(M-1)}^{(M)})$$
 is a

square symmetric matrix of order M x M, whose (i,j)-th element is $q_{ij}^{(M)} = C_{vi}^{(M)} C_{vj}^{(M)}$ (i,j=0,...,M-1). Since $C_{vu}^{(M)} = \pm 1$ and $(C_{vu}^{(S)})$ is non-singular, the linear equations in the matrices (L_{vu}) can be expressed in a form equivalent to (3.4) as,

(3.5)
$$\sum_{v=0}^{M-1} c_{vu}^{(M)} (L_{vu})(c^{(S)}) = I^{(S)}.$$

Since the unique solution to the equation $(H)(C^{(S)}) = I^{(S)}$ is $(H) = \frac{1}{S} (C^{(S)})'$, it follows that the M matrices $(L_{\mathbf{v}}^{(\mathbf{j})})$, $\mathbf{j} = 0, \ldots, M-1$, whose submatrices are given by:

(3.6)
$$(L_{vu}^{(j)}) = \begin{cases} \frac{1}{S} C_{vu}^{(M)} (C^{(S)}), & \text{if } u = j-1 \\ (0), & \text{otherwise} \end{cases}$$

constitute a basis of M independent solutions of (3.3). Thus, every g.l.s.e. (L_v) can be represented as a linear combination of the M linearly independent operators $(L_v^{(j)})$ $(j=0,\ldots,M-1)$, so that the coefficients of $(L_v^{(j)})$ add up to 1. Formally, the set of all g.l.s.e., given X_v is:

(3.7)
$$2(C_{v}) = \{(L_{v}) : (L_{v}) = \sum_{j=0}^{M-1} \lambda_{j} (L_{v}^{(j)}) ; \sum_{j=0}^{M-1} \lambda_{j} = 1\}$$

Every g.l.s.e. can thus be represented by M coordinates $(\lambda_0, \lambda_1, \ldots, \lambda_{M-1})$ such that $\sum_{j=0}^{M-1} \lambda_j = 1$. Furthermore, if $\hat{\beta}_V$ denotes the vector of g.l.s.e. of β then we have, according to (3.6) and (3.7)

(3.8)
$$\hat{\beta}_{v} = \frac{1}{S} \begin{bmatrix} \lambda_{o} & & & \\ \lambda_{1} & C_{vu}^{(M)} & & \\ \vdots & & & \\ \lambda_{M-1} & C_{vu}^{(M-1)} & & & \\ \end{bmatrix} \otimes (C^{(S)})^{v} Y(X_{v}) \text{ for every } v=0,...,M-1.$$

3.b. Some statistical properties of g.l.s.e.

In the present section we prove that there are no unbiased g.l.s.e. of β , and derive an expression for the trace of the mean-square-error matrix of a g.l.s.e. $\hat{\beta}$.

Consider a fractional replication design in which a block X_v $(v=0,\ldots,M-1)$ is chosen with probability ξ_v $(\xi_v\geq 0$ for all $v=0,\ldots,M-1$; $\sum_{v=0}^{M-1}\xi_v=1)$. A randomized fractional replication procedure is thus represented by an M-dimensional probability vector ξ . This class of randomization procedures contains, in particular, the fixed fractional replication design, in which one of the X_v blocks is chosen with probability one.

Theorem 1.

Let $\widetilde{\beta}^{(u)} = \frac{1}{S} C_{vu}^{(M)} (C^{(S)})$, $Y(X_v)$ then $E_{\xi *} \{\widetilde{\beta}^{(u)}\} = \beta^{(u)}$ for all $u=0,\ldots,M-1$ if, and only if, $\xi * = (\frac{1}{M},\frac{1}{M},\ldots,\frac{1}{M})$.

Proof.

The expected value of $\,\widetilde{\beta}^{(u)}\,\,$ under randomization procedure $\,\xi\,$, is given by:

Clearly, if $\xi_v = \frac{1}{M}$ for all v=0,...,M-1, then

$$\sum_{v=0}^{M-1} \xi_v C_{vu}^{(M)} C_{vw}^{(M)} = \frac{1}{M} \sum_{v=0}^{M-1} C_{vu}^{(M)} C_{vw}^{(M)} = 0$$

for all $u \neq w$ by the orthogonality of the column vectors of $(C^{(M)})$. Thus, $E_{\underline{\xi}*}(\widetilde{\beta}^{(u)}) = \beta^{(u)}$ for all $u=0,\ldots,M-1$. On the other hand, if $E_{\underline{\xi}}(\widetilde{\beta}^{(u)}) = \beta^{(u)}$ for all $u=0,\ldots,M-1$ then, in particular

(3.10)
$$E_{\xi}(\widetilde{\beta}^{(0)}) = \beta^{(0)} + \sum_{w=1}^{M-1} \sum_{v=0}^{M-1} \xi_{v} C_{vw}^{(M)} \beta^{(w)}$$

But the condition $\sum_{v=0}^{M-1} \xi_v C_{vw}^{(M)} = 0$ for all w=1,...,M-1 is equivalent to the condition:

$$(5.11) \qquad (c^{(M)}), \xi = \begin{pmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

Multiplying both sides of (3.11) by $(C^{(M)})$ we get:

(3.12)
$$M_{\xi} = (C^{(M)})(C^{(M)}); \quad \xi = (C^{(M)})\begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} = 1^{(M)}$$

where $1^{(M)}$ is an M-dimensional vector with unity in all its components. It follows that a necessary condition for the unbiasedness of $\widetilde{\beta}^{(u)}$ is that $\xi = \frac{1}{M} \ 1^{(M)}$, i.e., each block is chosen with the same probability. (Q.E.D.)

Returning to the g.l.s.e. we have:

(3.13)
$$E_{\xi *} \{ \hat{\beta} \} = E_{\xi *} \{ (\lambda_0 \tilde{\beta}^{(0)} : \lambda_1 \tilde{\beta}^{(1)} : ... : \lambda_{M-1} \tilde{\beta}^{(M-1)})' \}$$

$$= (\lambda_0 \beta^{(0)} : \lambda_1 \beta^{(1)} : ... : \lambda_{M-1} \beta^{(M-1)})' \}$$
where
$$\xi * = \frac{1}{M} \mathbf{1}^{(M)} = \frac{1}{M} (1,1,...,1)' .$$

Since $\sum_{u=0}^{M-1} \lambda_u = 1$ we conclude that there is no unbiased g.l.s.e. of β .

The g.l.s.e. in which $\lambda_0=1$ and $\lambda_u=0$ for all u>0 yields unbiased estimates of the components of $\beta^{(0)}$ only. Similarly when $\lambda_j=1$ $(j=0,\ldots,M-1)$ and $\lambda_j=0$ for all $j'\neq j$, the corresponding g.l.s.e. yields unbiased estimates of the components of $\beta^{(j)}$ only.

The mean-square-error dispersion matrix of a g.l.s.e. $\hat{\beta}$, under randomization procedure ξ , is defined by $E_{\xi}\{(\hat{\beta}-\beta)(\hat{\beta}-\beta)'\}$. Let $M(\xi,\lambda;\beta)$ denote the trace of the mean-square-error dispersion matrix of a g.l.s.e. represented by a vector λ , such that $\lambda'1^{(M)}=1$, under randomization procedure ξ ; i.e., $M(\xi,\lambda;\beta)=E_{\xi}\{(\hat{\beta}-\beta)'(\hat{\beta}-\beta)\}$.

Theorem 2.

The trace of the mean-square-error dispersion matrix under randomization procedure & is given by the expression:

$$(3.14) \qquad M(\xi,\lambda;\beta) = (\sigma^{2} + |\beta|^{2}) \sum_{u=0}^{M-1} \lambda_{u}^{2} - \sum_{u=0}^{M-1} (2\lambda_{u}^{-1}) |\beta^{(u)}|^{2} + \sum_{u_{1}=0}^{M-1} (2\lambda_{u}^{+1}) \sum_{u_{2}\neq u_{1}}^{M-1} \sum_{v=0}^{M-1} \left[\sum_{v=0}^{M-1} \xi_{v} C_{vu_{1}}^{(M)} C_{vu_{2}}^{(M)} \right] \beta^{(u_{1})'} \beta^{(u_{2})}$$

where $|\beta|^2 = \beta'\beta$, $|\beta^{(u)}|^2 = \beta^{(u)'}\beta^{(u)}$ (u=0,...,M-1).

Proof.

(3.15)
$$M(\xi,\lambda,\beta) = E_{\xi} \{(\hat{\beta}-\beta)'(\hat{\beta}-\beta)\} = E_{\xi} \{\hat{\beta}'\hat{\beta}\} - 2\beta' E_{\xi} \{\hat{\beta}\} + \beta'\beta$$

According to (3.8)

(3.16)
$$E_{\xi}(\hat{\beta}'\hat{\beta}) = \frac{1}{S} \sum_{u=0}^{M-1} \lambda_{u}^{2} E_{\xi}\{Y(X_{v})'Y(X_{v})\}$$

Substituting (2.19) for $Y(X_y)$ in (3.16) we get:

$$(3.17) \quad \frac{1}{S} \quad E_{\xi} \{Y(X_{v})^{!} \ Y(X_{v})\} = \frac{1}{S} E_{\xi} \{ \sum_{u_{1}=0}^{M-1} c_{vu_{1}}^{(M)} (c^{(S)}) \beta^{(u_{1})} + \epsilon \}^{!} \cdot \sum_{u_{2}=0}^{M-1} c_{vu_{2}}^{(M)} (c^{(S)}) \beta^{(u_{2})} + \epsilon \}^{!} =$$

$$= \sigma^{2} + \frac{1}{S} E_{\xi} \{ \sum_{u_{1}=0}^{M-1} \sum_{u_{2}=0}^{M-1} c_{vu_{1}}^{(M)} c_{vu_{2}}^{(M)} \beta^{(u_{1})} (c^{(S)})^{!} (c^{(S)}) \beta^{(u_{2})} \}$$

$$= \sigma^{2} + \sum_{u=0}^{M-1} |\beta^{(u)}|^{2} + \sum_{u_{1} \neq u_{2}} \sum_{v=0}^{M-1} (\sum_{v=0}^{M-1} \xi_{v} c_{vu_{1}}^{(M)} c_{vu_{2}}^{(M)} \beta^{(u_{1})} \beta^{(u_{2})} \beta^{(u_{2})}$$

Furthermore, $E_{\xi}(\hat{\beta}') = (\lambda_0 E_{\xi}(\tilde{\beta}^{(0)}') \dots \lambda_{M-1} E_{\xi}(\tilde{\beta}^{(M-1)'}))$ Substituting (3.9) for $E_{\xi}(\tilde{\beta}^{(u)})$, we arrive at

$$(3.18) \quad \beta' \quad E_{\xi}(\hat{\beta}) = \sum_{u=0}^{M-1} \lambda_{u} |\beta^{(u)}|^{2} + \sum_{u_{1}=0}^{M-1} \lambda_{u} \left[\sum_{u_{2} \neq u_{1}} \left(\sum_{v=0}^{M-1} \xi_{v} C_{vu_{1}}^{(M)} C_{vu_{2}}^{(M)} \right) \cdot \beta^{(u_{1})'} \beta^{(u_{2})} \right]$$

Thus, from (3.15)-(3.18) the result holds.

Q.E.D.

Corollary: When each block X_v (v=0,...,M-1) is chosen with equal probabilities ($\xi=\xi*$) we have

(3.19)
$$M(\xi^*,\lambda;\beta) = (\sigma^2 + |\beta|^2) \sum_{u=0}^{M-1} \lambda_u^2 - \sum_{u=0}^{M-1} (2\lambda_u^{-1}) |\beta^{(u)}|^2$$

3.c. Optimum strategies

A strategy of the Statistician is a pair of two M-dimensional vectors (ξ,λ) such that ξ is a probability vector, and λ' $1^{(M)}=1$. Every strategy (ξ,λ) represents a randomization procedure and a g.l.s.e. The decision problem is to choose (ξ,λ) optimally, with respect to the loss function $M(\xi,\lambda;\beta)$.

Comparing (3.14) to (3.19) it is easily verified that for every (ξ,λ) there exists β^0 in E^n such that $M(\xi,\lambda;\beta^0) > M(\xi^*,\lambda;\beta^0)$.

Thus, whenever β is arbitrary, ξ^* represents an admissible randomization procedure. For this reason we shall restrict the discussion from now on to strategies with randomization procedure ξ^* , and turn now to the problem of deciding upon an optimum g.l.s.e. under ξ^* .

We notice in (3.19) that $M(\xi^*,\lambda;\beta)$ depends on β only through the M values $|\beta^{(u)}|^2$. An a-priori information concerning these values might thus be utilized for the choice of λ . Thus, let $\Pi^{(u)}$ be an a-priori distribution of $|\beta^{(u)}|^2$, defined over the half-line $[0,\infty)$.

Theorem 3.

The Bayes g.l.s.e. of β , with respect to the a-priori distributions $\{\prod^{(0)},\ldots,\prod^{(M-1)}\}$, under randomization procedure ξ^* is determined by the vector $\lambda_\pi=(\lambda_\pi^{(0)},\ldots,\lambda_\pi^{(M-1)})$, where

(3.20)
$$\lambda_{\pi}^{(u)} = \frac{E_{\pi}(u)^{\{|\beta^{(u)}|^2\}}}{\sum_{u=0}^{M-1} E_{\pi}(u)^{\{|\beta^{(u)}|^2\}}}, \text{ for all } u=0,...,M-1.$$

Proof:

The risk function under (ξ^*,λ) and \prod is

(3.21)
$$R(\xi^*, \lambda; \Pi) = (\sigma^2 + \sum_{u=0}^{M-1} E_{u}(u) \{|\beta^{(u)}|^2\}) \sum_{u=0}^{M-1} \lambda_u^2 - \sum_{u=0}^{M-1} (2\lambda_u^{-1})$$

$$E_{\pi}(u)\{|\beta^{(u)}|^2\}$$
.

It is easily verified that $\lambda_{\pi}^{(u)}$ (u=0,...,M-1), given by (3.20), minimize (3.21) under the constraint $\sum_{u=0}^{M-1} \lambda_u = 1$.

Q.E.D.

Let
$$R_{\pi}^{(u)} = E_{\pi}(u)\{|\beta^{(u)}|^2\}$$
 (u=0,...,M-1) and $R_{\pi} = \sum_{u=0}^{M-1} R_{\pi}^{(u)}$ then the Bayes risk with respect to an a-priori distribution \prod is

(3.22)
$$R(\xi^*, \lambda_{\pi}; \Pi) = (\sigma^2 + R_{\pi}) \sum_{u=0}^{M-1} (\lambda_{\pi}^{(u)})^2 - \frac{M-1}{2} (2\lambda_{\pi}^{(u)} - 1) R_{\pi}^{(u)} = R_{\pi} - \frac{1}{R_{\pi}} (1 - \frac{\sigma^2}{R_{\pi}}) \sum_{u=0}^{M-1} (R_{\pi}^{(u)})^2$$

In particular, when all $|\beta^{(u)}|^2$ (u=0,...,M-l) have the same a-priori distribution, with $R_{\pi}^{(u)} = R_{\pi}^*$ for all u=0,...,M-l, then the Bayes g.l.s.e. is represented by $\lambda^* = (\frac{1}{M}, \frac{1}{M}, \ldots, \frac{1}{M})$ with a Bayes risk

(3.23)
$$R(\xi^*, \lambda^*, \prod) = \frac{\sigma^2}{M} + (M-1)R_{\pi}^*$$

Theorem 4.

 $\lambda^* = \frac{1}{M} \; 1^{\left(M\right)} \quad \text{represents the minimax and admissible g.l.s.e.}$ under randomization procedure $\; \xi^* \; \; \text{relative to the class of all}$ a-priori distributions $\; \prod \; , \; \text{such that} \; \; R_{\pi} = \sum_{u=0}^{M-1} R_{\pi}^{\left(u\right)} = \text{const. and}$ $(\sigma^2 < R_{\pi} < \infty)$. The minimax risk is given by (3.23).

Proof:

The minimax risk is the maximal Bayes risk, with respect to all the a-priori distributions \prod in the class considered. The Bayes risk for any of these π 's is given by (3.22) where $R_{\pi} = \sum_{u=0}^{M-1} R_{\pi}^{(u)}$ is a given constant. Set the Lagrangian

(3.24)
$$L(R_{\pi}^{(0)}, ..., R_{\pi}^{(M-1)}; \rho) = R_{\pi} - \frac{1}{R_{\pi}} (1 - \frac{\sigma^{2}}{R_{\pi}}) \sum_{u=0}^{M-1} (R_{\pi}^{(u)})^{2} + \rho(R_{\pi} - \sum_{u=0}^{M-1} R_{\pi}^{(u)})$$

By differentiating partially with respect to $R_{\pi}^{(u)}$ (u=0,...,M-1) and ρ and equating the derivatives to zero we arrive at the system of linear equations:

(3.25)
$$\begin{cases} -\frac{2}{R_{\pi}} (1 - \frac{\sigma^2}{R_{\pi}}) R_{\pi}^{(u)} = \rho & \text{for all } u=0,...,M-1 \\ \sum_{u=0}^{M-1} R_{\pi}^{(u)} = R_{\pi} \end{cases}$$

The solution of this system of linear equations is given by $R_{\pi}^{(u)} = \frac{R_{\pi}}{M}$ for every u=0,...,M-1. Furthermore, since $R_{\pi} \geq \sigma^2$, all the second order partial derivatives with respect to $R_{\pi}^{(u)}$ are negative. Thus all a-priori distributions π such that $R_{\pi}^{(u)} = \frac{R_{\pi}}{M}$ for every u=0,...,M-1 are minimax strategies for Nature. As mentioned before, $\lambda^* = \frac{1}{M} \, 1^{(M)}$ is then the unique minimax strategy for the Statistician. The Bayes risk corresponding to λ^* is given by (3.23). The admissibility of λ^* , relative to the class of a-priori distributions considered, follows from the fact that it is the unique minimax.

Q.E.D.

4. The generalized inverse and the g.l.s.e.

4.a. The g.l.s.e. of minimum norm

A. Ben-Israel and S. J. Wersan (1962) proved that the g.l.s.e. $(L_{\mathbf{V}}) \text{ with minimum norm, i.e., min tr } (L_{\mathbf{V}})'(I_{\mathbf{V}}) \text{ , is a particular } (L_{\mathbf{V}})$ generalized inverse $(C_{\mathbf{V}})^{\dagger}$ of the matrix of coefficients in (2.19), namely $(C_{\mathbf{V}}) = (1, C_{\mathbf{V}1}^{(M)}, \ldots, C_{\mathbf{V}(M-1)}^{(M)}) \times (C_{\mathbf{V}}^{(S)})$. The generalized inverse $(C_{\mathbf{V}})^{\dagger}$ always exists, it is unique, and given in general by the formula:

$$(C_{v})^{\dagger} = [I^{(N)} - (D_{v})(D_{v}^{\prime} D_{v})^{-1}(D_{v})^{\prime}](E_{v})(C_{v})^{\prime}$$

for all v=0,...,M-1; where (E_v) is a product of elementary transformations, which transforms $(C_v)'(C_v)$ into:

$$(4.2) \qquad (\mathbf{E}_{\mathbf{v}})(\mathbf{X}_{\mathbf{v}})(\mathbf{X}_{\mathbf{v}}) = \begin{bmatrix} \mathbf{I}^{(S)} & (\Delta_{\mathbf{v}}) \\ \dots & (\mathbf{v}) \end{bmatrix} \qquad (\mathbf{v} = 0, \dots, M-1)$$

and where

$$(4.3) \qquad (D_{\mathbf{v}}) = \begin{bmatrix} (\Delta_{\mathbf{v}}) \\ \dots \\ -T(N-S) \end{bmatrix} \qquad (v=0,\dots,M-1) .$$

The generalized inverse matrix, $(C_v)^{\dagger}$ has the properties:

(4.4)
$$(c_v)(c_v)^{\dagger}(c_v) = (c_v)$$
 for all v=0,...,M-1.

and

$$(\mathbf{c}_{\mathbf{v}})^{\dagger}(\mathbf{c}_{\mathbf{v}})(\mathbf{c}_{\mathbf{v}})' = (\mathbf{c}_{\mathbf{v}})'$$

A straightforward computation of $(C_v)^{\dagger}$ according to formula (4.1) yields the result

(4.5)
$$(c_{\mathbf{v}})^{\dagger} = \frac{1}{M} \begin{bmatrix} \tilde{\beta}^{(0)} \\ \tilde{\beta}^{(1)} \\ \vdots \\ \tilde{\beta}^{(M-1)} \end{bmatrix}$$

*

That is, $(C_v)^{\dagger}$ is a g.l.s.e. represented by $\lambda^* = \frac{1}{M} 1^{(M)}$, and has the optimal properties mentioned in the previous section. This result can be obtained in the present framework more easily. According to (3.8) a g.l.s.e. is given by

(4.6)
$$(L_{\mathbf{v}}) = \frac{1}{S} \begin{bmatrix} \lambda_{\mathbf{o}} & C_{\mathbf{vo}}^{(M)} \\ \vdots & & \\ \lambda_{M-1} & C_{\mathbf{v}(M-1)}^{(M)} \end{bmatrix} \otimes (C^{(S)}), \quad (\mathbf{v}=0,...,M-1)$$

Accordingly, the norm of (L,) is

(4.7)
$$\operatorname{tr.} (L_{\mathbf{v}})'(L_{\mathbf{v}}) = \operatorname{tr.} \left\{ \frac{1}{S^{2}} \sum_{u=0}^{M-1} \lambda_{u}^{2} (c^{(S)})(c^{(S)})' \right\}$$
$$= \frac{1}{S} \operatorname{tr.} \left\{ \sum_{u=0}^{M-1} \lambda_{u}^{2} \mathbf{I}^{(S)} \right\} = \sum_{u=0}^{M-1} \lambda_{u}^{2}.$$

Since the vector $\lambda^* = \frac{1}{M} \, 1^{(M)}$ minimizes $\sum_{u=0}^{M-1} \, \lambda_u^2$, under the constraint $\sum_{u=0}^{M-1} \, \lambda_u = 1$, it follows that the g.l.s.e. represented by λ^* minimizes the norm of (L_v) , $(v=0,\ldots,M-1)$.

4.b. The g.l.s.e. suggested by C. R. Rao.

C. R. Rao (1962) defines the g.l.s.e. of β by the operator $(L_v)^- = [(C_v)^!(C_v)]^- (C_v)^! (C_v)^! (C_v)^!$ is a generalized inverse of $(C_v)^!(C_v)$. In case $(C_v)^! (C_v)$ is invertible, $(L_v)^-$ is the

unique g.l.s.e. of β . We shall prove now that under the present model of fractional replication designs, Rao's g.l.s.e., $(L_{_{\rm V}})^-$, is represented by the vector $\lambda^-=(1,0,0,\ldots,0)$. For this purpose, define the matrix of elementary transformations

(4.8)
$$(\mathbf{E}_{\mathbf{v}}) = \sqrt{\frac{1}{\mathbf{S}}} \begin{bmatrix} \mathbf{1} & & \\ -\mathbf{C}_{\mathbf{v}_{1}}^{(M)} & & \\ \vdots & & \ddots \\ -\mathbf{C}_{\mathbf{v}_{(M-1)}}^{(M)} & & 1 \end{bmatrix} \mathbf{\otimes} \mathbf{I}^{(S)}$$

then we have, for every v=0,...,M-1,

(4.9)
$$(E_{\mathbf{v}})[(C_{\mathbf{v}})'(C_{\mathbf{v}})](E_{\mathbf{v}})' = \begin{bmatrix} I^{(S)} & (0) \\ - & 1 & - \\ (0) & (0) \end{bmatrix}$$

Hence,

(4.10)
$$[(C_v)'(C_v)]^* = (E_v)'(E_v)$$
.

To show this, consider the relationship

$$(4.11) \qquad [(C_{y})'(C_{y})][(C_{y})'(C_{y})]^{*}[(C_{y})'(C_{y})] = [(C_{y})'(C_{y})]$$

Multiply both sides of (4.11) from the left by $(E_{_{_{\bf V}}})$ and from the right by $(E_{_{_{\bf V}}})$ '. Then,

$$(4.12) \qquad (E_{v})[(C_{v})'(C_{v})](E_{v})'(E_{v})[(C_{v})'(C_{v})](E_{v})' = (E_{v})[(C_{v})'(C_{v})](E_{v})'$$

Or according to (4.9)

Accordingly,

$$= \frac{1}{S} \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \bigotimes (C^{(S)})' = \begin{bmatrix} \mathbf{\beta}^{(O)} \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

That is, Rao's g.l.s.e. $(L_v)^-$ is represented by $\lambda^- = (1,0,\ldots,0)$.

From theorem 1 it follows that $(L_v)^-$ is an unbiased estimator of $(\beta^{(0)},0)$. The trace of the dispersion matrix of $(L_v)^- Y(X_v)$, under randomization procedure §* is given according to (3.19) by

(4.15)
$$M(\xi^*, \lambda; \beta) = \sigma^2 + 2|\beta|^2 - 2|\beta^{(0)}|^2$$
$$= \sigma^2 + 2\sum_{u=1}^{M-1} |\beta^{(u)}|^2$$

Thus, in case all $|\beta^{(u)}|^2$ have the same a-priori distribution, the risk under strategies (ξ^*,λ^*) and π will be

(4.16)
$$R(\xi^*, \lambda^*, \pi) = \sigma^2 + 2(M-1) R_{\pi}^*$$

Comparing (4.16) to (3.23) we conclude that Rao's g.l.s.e. $(L_v)^-$ might be very far from the optimum g.l.s.e. in case all the subvectors of β have approximately the same average effect. On the other hand, in case the effects of $\beta^{(1)}$ to $\beta^{(M-1)}$ are negligible relative to the effect of $\beta^{(0)}$ the Bayes g.l.s.e. will be very close to Rao's g.l.s.e.

Acknowledgment

The author gratefully acknowledges Professors H. Solomon and H. Chernoff for their helpful comments and encouragement.

References

- 1. A. P. Dempster (1960), "Random allocation designs I: On general classes of estimation methods", Ann. Math. Stat. Vol. 31, pp. 885-905.
- 2. A. P. Dempster (1961), "Random allocation designs II: Approximative theory for simple random allocations", Ann. Math. Stat. Vol. 32, pp. 387-405.
- 3. S. Ehrenfeld and S. Zacks (1961), "Randomization and factorial experiments", Ann. Math. Stat. Vol. 32, pp. 270-297.
- 4. S. Ehrenfeld and S. Zacks (1962), "Optimum strategies in factorial experiments", submitted for publication in the Ann. Math. Stat.
- 5. O. Kempthorne, The Design and Analysis of Experiments, John Wiley, New York (1952).
- 6. C.R. Rao (1962), "A note on a generalized inverse", <u>Jour. Roy. Stat. Soc.</u> (B) Vol. 24, p. 152.
- 7. B.V. Shah and O. Kempthoren (1962a), "Some properties of random allocation designs", <u>Technical Report</u>, Stat. Lab. Iowa Stat. Univ.
- 8. B.V. Shah and O. Kempthorne (1962b), "Randomization in fractional factorial", Technical Report, Stat. Lab. Iowa State Univ.
- 9. K. Takeuchi (1961), "On a special class of regression problems and its applications: Random Combined Fractional Factorial Designs", Rep. Stat. Appl. Res. JUSE, Vol. 7, No. 4, pp. 1-33.
- 10. K. Takeuchi (1961), "On a special class of regression problems and its application: some remarks about general models", Rep. Stat. Appl. Res. JUSE. Vol. 8, No. 1, pp. 7-17.
- 11. S. Zacks (1962), "On a complete class of linear unbiased estimators for randomized factorial experiments", submitted for publication in the Ann. Math. Stat.

STANFORD UNIVERSITY TECHNICAL REPORTS DISTRIBUTION LIST CONTRACT Non-225(52)

Armed Services Technical		Commanding Officer		Document Library	
Information Agency		Frankford Arsenal		U.S. Atomic Energy Commission	
Arlington Hail Station	10	Library Branch, 0270, Bidg. 40		19th and Constitution Aves. N.W. Washington 25, D. C.	1
Arlington 12, Virginia	10	Bridge and Tacony Streets Philadelphia 37, Pennsylvania	1	Washington 25, D. C.	-
Bureau of Supplies and Accounts		rittadelpita 37, remisyrtems	•	Headquarters	
Code OW		Commanding Officer		Oklahoma City Air Materiel Area United States Air Force	
Department of the Navy	_	Rock Island Arsenal Rock Island, Illinois	1	United States Air Force Tinker Air Ferce Base,	
Washington 25, D C.	1	Reck Island, Illinois		Oklahema	1
the delication and Mathematical		Commanding General			
Head, Logistics and Mathematical Statistics Branch	•	Redstone Arsenal (ORDD\Y-QC)	_	Institute of Statistics	
Office of Naval Research		Huntsville, Alabama	1	Institute of Statistics North Carolina State College of A & E Raleigh, North Carolina	1
Cade 436	3	Commanding General		Katelan' Mount Cerotine	•
Washington 25, D. C.	,	White Sands Proving Ground		Jet Propulsion Laboratory	
Commanding Officer		White Sands Proving Ground (ORDBS-TS-TIB)		California Institute of Technology	
Office of Naval Research		Las Cruces, New Mexico	1	Attn: A.J. Stosick 4800 Oak Grove Drive	
Branch Office		Commanding General		Pasadena 3, California	1
Navy No. 100, Floot P. O. New York, N. Y.	2	Attn: Paul C. Cox, Ord. Mission			
New York, W. Y.	-	White Sands Proving Ground		Librarian	
Commanding Officer		Las Cruces, New Mexico	1	The RAND Corporation	
Diffice of Mayal Research		Companding General		1700 Main Street Santa Menica, California	1
Branch Office 1000 Geary Street		Commanding General Attn: Technical Documents Center			
San Francisco 9, California	1	Signal Corps Engineering Laboratory Fort Menmouth, New Jersey		Library Division Navel Missile Center Command U.S. Navel Missile Center	
		Fort Menmouth, New Jersey	1	HAVE MISSHE CERRER COMMEND	
Commanding Officer		Commandina General		Attn: J. L. Nickel	
Office of Naval Research Branch Office		Commanding General Ordnance Weapons Command Attn: Research Branch		Attn: J. L. Nickel Point Mugu, California	1
1 Och Floor . The John Crerar		Attn: Research Branch		Makasasa Bhilalan	
10th Floor, The John Crerar Library Bidg		Rock Island, Illinois	1	Mathematics Division Code 5077	
86 East Randolph Street Chicago 1, Ilfinois	1	Commanding General		U.S. Naval Ordnance Test Station	
Chicago 1, Illinois	•	U.S. Army Electronic Proving Ground		China Lake, California	1
Commanding Officer		Fort Huachuca, Arizena Attn: Technical Library		NACA	
Ultice of Mayal Research		Attn: Technical Library	1	NASA Attn: Mr. E.B. Jackson, Office	
Branch Office		Commander		of Aero Intelligence	
346 Braidway New York 13, N. Y.	1	Wright Air Development Center		1724 F Street, N. W. Washington 25, D. C.	1
Hem 10x 23, 71. 7.		Attn: ARL Tech. Library, WCRR	,	Washington 25, D. C.	
Commanding Officer		Wright-Patterson Air Force Base, Ohio	1	National Applied Mathematics Labs.	
Diamoné Ordance Fuze Labs. Washington 25, D. C.	1	Commander		National Applied Mathematics Labs, National Bureau of Standards Washington 25, D. C.	_
Washington 20, U. C.	-	Western Development Division, WDSIT		Washington 25, D. C.	1
Commanding Officer		P.O. Box 262 Inglewood, California	1	Naval Inspector of Ordnance	
Picatinny Arsenal (ORDBB-TH8) Dever, New Jersey	1	Inglewood, California	•	II S Mayal Gun Factory	
Dover, New Jersey	•	Chief, Research Division		Washington 25, D. C. Attn: Mrs. C. D. Hock	
Commanding Officer		Office of Research & Development		Attn: Mrs. C. D. Hock	1
Waterlown Arsenal (OMRO) Waterlown 72, Massachusetts	1	Office of Chief of Staff U.S. Army		Office, Asst. Chief of Staff, G-4	
Waterlown 72, Massachusetts		Washington 25, D. C.	1	Research Branch, R & D Division	
Commendate Officer				Department of the Army Washington 25, D. C.	1
Commanding Officer Atin: W. A. Labs Watertown Arsenal		Chief, Computing Laboratory		Washington 25, D. C.	•
Watertown Arsenal	1	Ballistic Research Laboratory Aberdeen Proving Ground, Maryland	1	Superintendent	
Watertown 72, Massachusetts	•	Aperdeen Froming Ground, many tank	_	ti S. Navy Postgraduate School	
Commanding Officer		Director		Monterey, California	1
Watervilet Arsenal		National Security Agency Attn: REMP-1		Attn: Library	•
Watervilet, New York	1	Attn: NEMP-1 Fort George G. Meade, Maryland	2	Technical Information Officer	
Commanding Officer				Naval Research Laboratory	
Attn. Inspection Division		Director of Operations		Washington 25, D. C.	6
Springfield Armory		Operations Analysis Div., AFOOP		Technical information Service	
Springfield, Massachusetts	1	Hq., U.S. Air Force Washington 25, D. C.	1	Attn: Reference Branch P.O. Box 62	
Commanding Officer		trainington as you a.		P.O. Box 62	1
Signal Cores Electronic Research		Director		Oak Ridge, Tennessee	
Signal Corps Electronic Research Unit, EDL		Snow, Ice & Permafrost Research Establishment		Technical Library Branch	
9560 Technical Service Unit		Corps of Engineers		Code 234	
P.O. Box 205 Mountain View, California	1	Corps of Engineers 1215 Washington Avenue	,	II C Neval Ordnance Laboratory	
		Wilmette, Illinois	1	Attn: Clayborn Graves Corona, California	1
Commanding Officer		Director		coone, celimine	-
		Lincoln Laboratory		Institute for Defense Analyses	
Willow Run Research Certier		Lexington, Massachusetts	1	Communications Research Division	
Army Llaison Group, Project Michigan Willow Run Research Center Ypsilanti, Michigan	1			von Neumann Hall Princeton, New Jersey	1
		Department of Mathematics Michigan State University		Franceion, new Jersey	•
Commanding Officer Engineering Research & Development Lab	١.	East Lansing, Michigan	1		
Fort Belvoir, Virginia	1				
				A 106.2	

Mr. Irving B. Altman Inspection & QC Division Office, Asst. Secretary of Defense Room 28670, The Pentages Washington 25, D. C.	1	Professor Solomon Kullback Department of Statistics George Washington University Washington 7, D. C.	1	Professer L. J. Savege Mathematics Department University of Michigan Ann Arber, Michigan	1
Professor T. W. Anderson Department of Statistics Columbia University New York 27, New York	1	Professor W. H. Kruskal Department of Statistics The University of Chicago Chicago, Illinois	1	Prefesser W. L. Smith Statistics Department University of North Carolina Chapel Hill , North Carolina	1
Professor Robert Bechinder Dept. of Industrial and Engineering Administration Sibley School of Mechanical Engineering Cemell University Itheca, New York		Professor Eugene Lukacs Department of Mathematics Catholic University Washington 15, D. C. Dr. Craig Magwire	1	Dr. Milton Sobel Statistics Department University of Minneseta Minneapolis, Minneseta Md. G. R. Stand	1
	1	2954 Winchester Way Rancho Cordova, California	1	Mr. G. P. Steck Division 5511 Sandia Carp., Sandia Bose Albuquerque, New Mexico	1
Professor Fred. C. Andrews Department of Mathematics University of Oregon Eugene, Grugon	1	Professor G. W. McEirath Department of Mechanical Engineering University of Minnesota Minneapolis 14, Minnesota	1	Professor Donald Truss Department of Mathematics University of Oregon Eugene , Oregon	
Professor Z. W. Blimbaum Department of Mathematics University of Washington Seattle 5, Washington	1	Dr. Knox T. Milisaps Executive Director Air Force Office of Scientific Research Washington 25, D. C.	1	Professor John W. Tukey Department of Mathematics Princeton University	1
Dr. David Blackwell Department of Mathematical Sciences University of California Berkeley 4, California	1	D. E. Newnham Chief, Ind. Engr. Div. Comptroller Hq., San Bernardino Air Materiel Area USAF, Norton Air Force Base, California		Princeton, New Jersey Professor G. S. Wetson Department of Mathematics University of Terente, Terento 5, Ontario, Canada	1
Professor Raiph A. Bradley Department of Statistics Florida State University		Professor Edwin G. Olds	1		1
Taliahassee, Florida Dr. John W. Cell Drandment of Mathematics	1	College of Engineering and Sciences Carnegle Institute of Technology Pittsburgh 13, Pennsylvania	ı	Dr. Harry Weingarten Special Projects Office, SP2016 Navy Department Washington 25, D. C.	1
North Carolina State College Raleigh, North Carolina Professor William G. Cochran	1	Dr. William R. Pabst Bureau of Weapons Room 0306, Main Navy Department of the Navy Washington 25, D. C.		Dr. F. J. Weyl, Director Mathematical Sciences Division Office of Naval Research Washington 25, D. C.	1
Department of Statistics Harvard University 2 Divinity Avenue, Room 311 Cambridge 38, Massachusetts	1	Washington 25, D. C. Mr. Edward Paulson * 72-10 41 Ave. Woodside 77	1	Dr. John Wilkes Office of Naval Research, Code 200 Washington 25, D. C.	1
Miss Besse B Day Bureau of Ships, Code 34 20 Room 3 210 Main Navy Department of the Navy Washington 25, D. C.	-	New York , New York	1	Professor S. S. Wilks Department of Mathematics Princeton University	•
	1	Reliability Branch, 750 Diamond Ordnance Fuze Laboratory Room 105, Building 83 Washington 25, D. C	1	Princeton, New Jersey	1
Dr. Walter L. Deemer, Jr. Operations Analysis Div., DCE/O Hq., U.S. Air Force Washington 25, D. C.	1	Professor Ronald Pyke Mathematics Department	•	Mr. Silas Williams Standards Branch, Proc. Div. Office, DC/S for Legistics Department of the Army Washington 25, D. C.	1
Professor Cyrus Derman Dept, of Industrial Engineering Columbia University New York 27, New York	1	University of Washington Seattle 5, Washington Dr. Paul Rider Wilchi & Jr. Davelorment Center, MCPRM	1	Professor Jacob Wolfowitz Department of Mathematics Cornell University Ithaca, New York	1
Dr. Donald P. Gaver Westinghouse Research Labs . Beufah Rd . Churchill Boro . Pittsburgh 35 , Pa .	•	Wright Air Development Center, WCRRM Wright-Patterson A.F.B., Ohio Professor Herbert Robbins	1	Mr. William W. Wolman Code MER - Bidg. T-2 Room C301 700 Jackson Piace, N. W. Washington 25, D. C.	•
No. Manuald Combat	1	Professor Herbert Robbins Dept. of Mathematical Statistics Columbia University New York 27, New York Professor Murray Rosenblatt	1		1
Mr. narvid Genetics Head, Operations Research Group Code 01-2 Pacific Missile Range Box 1	1	Department of Mathematics Brown University Providence 12, Rhode Island	1	Marvin Zelen Mathematics Research Center U. S. Army University of Wisconsin Madison b , Wisconsin	1
Point Mugu, California Dr. Ivan Hershner Office, Chief of Research & Dev. U.S. Army, Research Division 3E382 Washington 25, D. C.	1	Professor Herman Rubin Department of Statistics Michigan State University East Lansing, Michigan	1	Additional copies for project leader and assistants and reserve for future requirements	50
Professor W Hirsch Institute of Mathematical Sciences New York University New York 3, New York	1	Professor J. S. Rusuagi College of Medicine University of Cincinnati Cincinnati, Ohjo	1		
Mr. Eugene Hixson Code 600.1 GSFC, NASA Greenbelt, Maryland	1	Professor I. R. Savage School of Business Administration University of Minnesota Minneapolis, Minnesota	1		
Professor Harold Hotelling Department of Statistics University of North Carolina Chapel Hill, North Carolina	1	Miss Marton M. Sandomire 2281 Cedar Street Berkeley 9, California	1		

JOINT SERVICES ADVISORY GROUP

Mr. Fred Frishman Army Research Office Arlington Hall Station Arlington, Virginia	1	Lt. Col. John W. Querry, Chief Applied Mathematics Division Air Force Office of Scientific Research Washington 25, D. C.
Mrs. Dorothy M. Gilford Mathematical Sciences Division Office of Naval Research Washington 25, D. C.	3	Major Oliver A. Shaw, Jr. Mathematics Division Air Force Office of Scientific Research Washington 25, D. C. 2
Dr. Robert Lundegard Logistics and Mathematical Statistics Branch Office of Naval Research Washington 25, D. C.	1	Mr. Carl L. Schaniel Code 122 U.S. Naval Ordnance Test Station China Lake, California
Mr. R. H. Noyes Inst. for Exploratory Research USASRDL Fort Monmouth, New Jersey	1	Mr. J. Weinstein Institute for Exploratory Research USASRDL Fort Monmouth, New Jersey 1